

## Task-2 : Stock Market Prediction And Forecasting Using Stacked LSTM

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```
In [1]: import math
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
import pandas_datareader as pdr
import tensorflow as tf
```

```
In [2]: url='https://raw.githubusercontent.com/mwitiderrick/stockprice/master/NSE-TATAGLOBAL.csv'
data=pd.read_csv(url)
```

```
In [3]: data.head()
```

```
Out[3]:
```

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
0	2018-09-28	234.05	235.95	230.20	233.50	233.75	3069914	7162.35
1	2018-09-27	234.55	236.80	231.10	233.80	233.25	5082859	11859.95
2	2018-09-26	240.00	240.00	232.50	235.00	234.25	2240909	5248.60
3	2018-09-25	233.30	236.75	232.00	236.25	236.10	2349368	5503.90
4	2018-09-24	233.55	239.20	230.75	234.00	233.30	3423509	7999.55

```
In [4]: data.shape
```

```
Out[4]: (2035, 8)
```

```
In [5]: data.tail()
```

```
Out[5]:
```

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
<b>2030</b>	2010-07-27	117.6	119.50	112.00	118.80	118.65	586100	694.98
<b>2031</b>	2010-07-26	120.1	121.00	117.10	117.10	117.60	658440	780.01
<b>2032</b>	2010-07-23	121.8	121.95	120.25	120.35	120.65	281312	340.31
<b>2033</b>	2010-07-22	120.3	122.00	120.25	120.75	120.90	293312	355.17
<b>2034</b>	2010-07-21	122.1	123.00	121.05	121.10	121.55	658666	803.56

```
In [6]: df=data.sort_values(by="Date")
df.head()
```

```
Out[6]:
```

	Date	Open	High	Low	Last	Close	Total Trade Quantity	Turnover (Lacs)
<b>2034</b>	2010-07-21	122.1	123.00	121.05	121.10	121.55	658666	803.56
<b>2033</b>	2010-07-22	120.3	122.00	120.25	120.75	120.90	293312	355.17
<b>2032</b>	2010-07-23	121.8	121.95	120.25	120.35	120.65	281312	340.31
<b>2031</b>	2010-07-26	120.1	121.00	117.10	117.10	117.60	658440	780.01
<b>2030</b>	2010-07-27	117.6	119.50	112.00	118.80	118.65	586100	694.98

```
In [7]: df=df.reset_index().Close
df
```

```
Out[7]:
```

0	121.55
1	120.90
2	120.65
3	117.60
4	118.65
	...
2030	233.30
2031	236.10

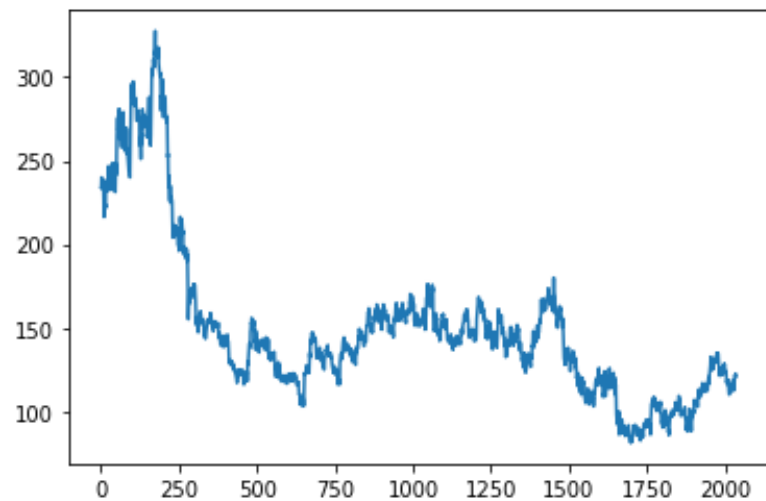
```
2032    234.25
2033    233.25
2034    233.75
Name: Close, Length: 2035, dtype: float64
```

```
In [8]: dff=data.Open
dff
```

```
Out[8]: 0      234.05
1      234.55
2      240.00
3      233.30
4      233.55
...
2030    117.60
2031    120.10
2032    121.80
2033    120.30
2034    122.10
Name: Open, Length: 2035, dtype: float64
```

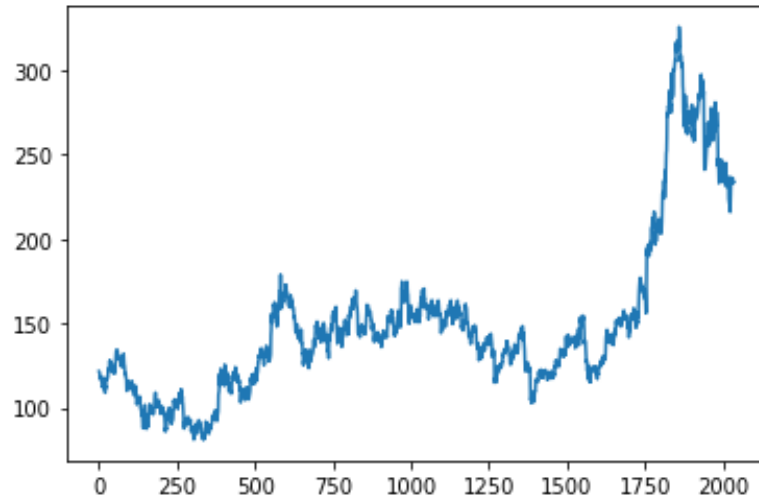
```
In [9]: dff.plot()
```

```
Out[9]: <AxesSubplot:>
```



In [10]: `df.plot()`

Out[10]: `<AxesSubplot:>`



In [11]: `#Transform features by scaling each feature to a given range`  
`from sklearn.preprocessing import MinMaxScaler`

```
sc = MinMaxScaler(feature_range = (0, 1))  
df = sc.fit_transform(np.array(df).reshape(-1,1))
```

In [27]: `df`

Out[27]: `array([[0.16584967],  
 [0.16319444],  
 [0.1621732 ],  
 ...,  
 [0.62622549],  
 [0.62214052],  
 [0.62418301]])`

**starting with train,test set for training and testing**

```
In [12]: train_size = int(len(df) * 0.70)

test_size = len(df) - train_size

trainSet, testSet = df[0:train_size, :], df[train_size:len(df), :1]
```

## Data set Generation

```
In [13]: def datasetGen(data, length):
        arA, arB = [], []
        for i in range(len(data)-length-1):
            arA.append(data[i:(i + length), 0])
            arB.append(data[i + length, 0])
        return np.array(arr1), np.array(arr2)
```

```
In [14]: X_train, Y_train = datasetGen(trainSet, 100)
        X_test, Y_test = datasetGen(testSet, 100)
```

```
In [28]: print(X_train.shape)
        print()
        X_train
```

(1323, 100, 1)

```
Out[28]: array([[0.16584967],
               [0.16319444],
               [0.1621732 ],
               ...,
               [0.14011438],
               [0.13848039],
               [0.12479575]],

              [[0.16319444],
               [0.1621732 ],
               [0.14971405],
               ...,
               [0.13848039],
```

```

[0.12479575],
[0.12254902]],

[[0.1621732 ],
 [0.14971405],
 [0.15400327],
 ...,
 [0.12479575],
 [0.12254902],
 [0.13010621]],

...,

[[0.18280229],
 [0.19178922],
 [0.19219771],
 ...,
 [0.16441993],
 [0.15236928],
 [0.15482026]],

[[0.19178922],
 [0.19219771],
 [0.19281046],
 ...,
 [0.15236928],
 [0.15482026],
 [0.15420752]],

[[0.19219771],
 [0.19281046],
 [0.21058007],
 ...,
 [0.15482026],
 [0.15420752],
 [0.15012255]]])

```

In [31]:

```

print(Y_train.shape)
print()
Y_train

```

```
(1323,)
```

```
Out[31]: array([0.12254902, 0.13010621, 0.13541667, ..., 0.15420752, 0.15012255,  
              0.15747549])
```

```
In [32]: print(X_test.shape)  
print()  
X_test
```

```
(510, 100, 1)
```

```
Out[32]: array([[0.1621732 ],  
                [0.15951797],  
                [0.16707516],  
                ...,  
                [0.24775327],  
                [0.24836601],  
                [0.25020425]],  
               [[0.15951797],  
                [0.16707516],  
                [0.15992647],  
                ...,  
                [0.24836601],  
                [0.25020425],  
                [0.25020425]],  
               [[0.16707516],  
                [0.15992647],  
                [0.16053922],  
                ...,  
                [0.25020425],  
                [0.25020425],  
                [0.23999183]],  
               ...,  
               [[0.86519608],  
                [0.84027778],  
                [0.84497549],  
                ...],
```

```

[0.62888072],
[0.62765523],
[0.62234477]],

[[0.84027778],
[0.84497549],
[0.87111928],
...,
[0.62765523],
[0.62234477],
[0.63378268]],

[[0.84497549],
[0.87111928],
[0.85273693],
...,
[0.62234477],
[0.63378268],
[0.62622549]]])

```

In [33]:

```

print(Y_test.shape)
print()
Y_test

```

(510,)

Out[33]:

```

array([0.25020425, 0.23999183, 0.22201797, 0.23080065, 0.22896242,
       0.24060458, 0.24080882, 0.23304739, 0.23325163, 0.23672386,
       0.25102124, 0.24387255, 0.25980392, 0.26879085, 0.22896242,
       0.23958333, 0.25081699, 0.24857026, 0.29554739, 0.28574346,
       0.2871732 , 0.28880719, 0.27328431, 0.29513889, 0.28308824,
       0.28982843, 0.30085784, 0.29166667, 0.29227941, 0.29861111,
       0.28206699, 0.26327614, 0.23202614, 0.24101307, 0.23876634,
       0.23549837, 0.22058824, 0.20588235, 0.1875      , 0.19219771,
       0.18995098, 0.16727941, 0.1815768 , 0.15400327, 0.14726307,
       0.16482843, 0.15379902, 0.15992647, 0.14174837, 0.14848856,
       0.1499183 , 0.13950163, 0.15339052, 0.16339869, 0.16748366,
       0.1752451 , 0.16727941, 0.16584967, 0.1744281 , 0.17340686,
       0.1689134 , 0.1752451 , 0.17544935, 0.17320261, 0.17320261,
       0.17075163, 0.17769608, 0.17708333, 0.17708333, 0.16462418,
       0.16605392, 0.15706699, 0.15482026, 0.14726307, 0.16176471,

```



0.15788399, 0.16401144, 0.16768791, 0.17401961, 0.17769608,  
0.17769608, 0.1807598 , 0.17851307, 0.17892157, 0.17769608,  
0.19035948, 0.18055556, 0.19362745, 0.19076797, 0.19056373,  
0.19485294, 0.20241013, 0.18566176, 0.19178922, 0.19403595,  
0.19362745, 0.19709967, 0.20220588, 0.19566993, 0.20077614,  
0.23161765, 0.24877451, 0.25245098, 0.24795752, 0.26633987,  
0.25714869, 0.25735294, 0.24918301, 0.24611928, 0.23611111,  
0.24162582, 0.24183007, 0.24775327, 0.24775327, 0.24121732,  
0.23815359, 0.23672386, 0.23325163, 0.2559232 , 0.24571078,  
0.25490196, 0.25449346, 0.25245098, 0.24693627, 0.24448529,  
0.22937092, 0.23917484, 0.24489379, 0.26041667, 0.26082516,  
0.26756536, 0.26531863, 0.25388072, 0.28860294, 0.28676471,  
0.27634804, 0.28431373, 0.28451797, 0.28267974, 0.28431373,  
0.28982843, 0.29289216, 0.29452614, 0.28451797, 0.28594771,  
0.29595588, 0.28921569, 0.28410948, 0.27982026, 0.2745098 ,  
0.28288399, 0.29084967, 0.28737745, 0.29003268, 0.28901144,  
0.29738562, 0.29064542, 0.29473039, 0.31352124, 0.30739379,  
0.30392157, 0.29146242, 0.29268791, 0.29861111, 0.30187908,  
0.29840686, 0.29289216, 0.30085784, 0.3002451 , 0.29370915,  
0.27185458, 0.28002451, 0.27308007, 0.26062092, 0.24816176,  
0.25367647, 0.26940359, 0.26327614, 0.26388889, 0.27961601,  
0.29187092, 0.30187908, 0.30228758, 0.2943219 , 0.30473856,  
0.30085784, 0.30004085, 0.29064542, 0.30841503, 0.31658497,  
0.3192402 , 0.31556373, 0.30821078, 0.31004902, 0.30984477,  
0.30147059, 0.28676471, 0.27124183, 0.27859477, 0.27593954,  
0.28472222, 0.28860294, 0.28227124, 0.30718954, 0.29575163,  
0.31372549, 0.3496732 , 0.36172386, 0.37908497, 0.38582516,  
0.3809232 , 0.39236111, 0.37908497, 0.38480392, 0.3690768 ,  
0.37275327, 0.36376634, 0.36846405, 0.37275327, 0.35723039,  
0.36029412, 0.3619281 , 0.35355392, 0.34375 , 0.3500817 ,  
0.34926471, 0.36560458, 0.34742647, 0.3314951 , 0.3065768 ,  
0.34007353, 0.3995098 , 0.46119281, 0.44791667, 0.46343954,  
0.47058824, 0.44485294, 0.47222222, 0.4628268 , 0.45894608,  
0.44689542, 0.47426471, 0.47814542, 0.47651144, 0.45894608,  
0.47283497, 0.51062092, 0.50980392, 0.51450163, 0.52410131,  
0.53553922, 0.51613562, 0.53553922, 0.52941176, 0.54064542,  
0.55249183, 0.54227941, 0.53615196, 0.49101307, 0.47263072,  
0.50531046, 0.48059641, 0.49918301, 0.51000817, 0.5126634 ,  
0.50653595, 0.5249183 , 0.52553105, 0.52022059, 0.52083333,  
0.52471405, 0.53533497, 0.52818627, 0.52553105, 0.52839052,  
0.51409314, 0.50510621, 0.49897876, 0.52573529, 0.50796569,  
0.52634804, 0.57005719, 0.59579248, 0.59640523, 0.59987745,

0.59395425, 0.58966503, 0.62806373, 0.58843954, 0.58639706,  
0.65155229, 0.63398693, 0.62683824, 0.67708333, 0.6942402 ,  
0.69852941, 0.7310049 , 0.79227941, 0.78513072, 0.78880719,  
0.79187092, 0.80780229, 0.80269608, 0.84477124, 0.81392974,  
0.84211601, 0.82761438, 0.80923203, 0.8129085 , 0.79248366,  
0.82945261, 0.84763072, 0.8880719 , 0.84497549, 0.83129085,  
0.8306781 , 0.84109477, 0.85743464, 0.89746732, 0.89828431,  
0.89542484, 0.90604575, 0.91911765, 0.91748366, 0.92177288,  
0.96180556, 0.94750817, 0.94035948, 0.94056373, 0.94485294,  
0.94771242, 0.96670752, 0.95731209, 0.95241013, 0.94914216,  
1. , 0.98876634, 0.96180556, 0.9501634 , 0.91707516,  
0.92892157, 0.93096405, 0.92422386, 0.9121732 , 0.9121732 ,  
0.90216503, 0.91115196, 0.85661765, 0.84722222, 0.79207516,  
0.78492647, 0.75878268, 0.81372549, 0.81495098, 0.80596405,  
0.83394608, 0.82107843, 0.80923203, 0.79513889, 0.76388889,  
0.74305556, 0.75837418, 0.74019608, 0.77512255, 0.78656046,  
0.79268791, 0.79473039, 0.79207516, 0.79534314, 0.78921569,  
0.76776961, 0.78615196, 0.76552288, 0.79697712, 0.8057598 ,  
0.80698529, 0.81147876, 0.78206699, 0.73386438, 0.73978758,  
0.74448529, 0.72671569, 0.72283497, 0.74040033, 0.74693627,  
0.72630719, 0.80065359, 0.79187092, 0.77328431, 0.78451797,  
0.78329248, 0.80412582, 0.79473039, 0.7879902 , 0.7879902 ,  
0.79473039, 0.81168301, 0.81699346, 0.82986111, 0.83843954,  
0.83884804, 0.82781863, 0.82598039, 0.85028595, 0.85886438,  
0.86254085, 0.88480392, 0.86519608, 0.84027778, 0.84497549,  
0.87111928, 0.85273693, 0.86213235, 0.83455882, 0.84375 ,  
0.71466503, 0.66053922, 0.65420752, 0.68035131, 0.6689134 ,  
0.66421569, 0.67892157, 0.69138072, 0.70894608, 0.72263072,  
0.73406863, 0.72120098, 0.74979575, 0.7689951 , 0.74468954,  
0.7120098 , 0.74673203, 0.75490196, 0.76041667, 0.75980392,  
0.74652778, 0.76245915, 0.77062908, 0.80412582, 0.78104575,  
0.79411765, 0.76021242, 0.73488562, 0.72406046, 0.73999183,  
0.74979575, 0.76143791, 0.75428922, 0.74754902, 0.76960784,  
0.7495915 , 0.81719771, 0.80208333, 0.77736928, 0.75714869,  
0.79370915, 0.79166667, 0.77144608, 0.75265523, 0.72691993,  
0.6629902 , 0.68198529, 0.64419935, 0.62152778, 0.6439951 ,  
0.66319444, 0.67544935, 0.63705065, 0.64603758, 0.65890523,  
0.65870098, 0.67790033, 0.63194444, 0.63112745, 0.62785948,  
0.64767157, 0.63991013, 0.63480392, 0.63868464, 0.62336601,  
0.62745098, 0.63970588, 0.64705882, 0.66196895, 0.67075163,  
0.65196078, 0.63112745, 0.61294935, 0.6376634 , 0.62254902,  
0.62745098, 0.63337418, 0.62642974, 0.63623366, 0.58312908,

```
0.57781863, 0.57230392, 0.58006536, 0.57618464, 0.55167484,  
0.57883987, 0.625      , 0.63582516, 0.62949346, 0.62888072,  
0.62765523, 0.62234477, 0.63378268, 0.62622549, 0.62214052])
```

## reording Column Inputs

```
In [34]: X_train = X_train.reshape(X_train.shape[0], X_train.shape[1], 1)  
X_test  = X_test.reshape(X_test.shape[0], X_test.shape[1], 1)
```

## Starting with Stacked LSTM model

```
In [35]: from keras.models import Sequential  
from keras.layers import Dense, LSTM
```

```
In [40]: model = Sequential()  
  
## Stacked LSTM model with 50 hidden layers.  
model.add(LSTM(50, return_sequences= True, input_shape= (100,1)))  
model.add(LSTM(50, return_sequences= True))  
model.add(LSTM(50, return_sequences= True))  
model.add(LSTM(50))  
  
## Output Layer  
model.add(Dense(1))  
  
## Compile the model  
model.compile(loss= 'mean_squared_error', optimizer= 'adam')
```

```
In [41]: model.summary()
```

Model: "sequential\_2"

Layer (type)	Output Shape	Param #
=====		
lstm_8 (LSTM)	(None, 100, 50)	10400

lstm_9 (LSTM)	(None, 100, 50)	20200
lstm_10 (LSTM)	(None, 100, 50)	20200
lstm_11 (LSTM)	(None, 50)	20200
dense_2 (Dense)	(None, 1)	51

```

=====
Total params: 71,051
Trainable params: 71,051
Non-trainable params: 0

```

---

In [42]:

```

## Training Procedure
model.fit(X_train, Y_train, validation_data= (X_test, Y_test), epochs= 100, batch_size= 64, verbose= 1)

```

```

Epoch 1/100
21/21 [=====] - 26s 628ms/step - loss: 0.0083 - val_loss: 0.0241
Epoch 2/100
21/21 [=====] - 9s 447ms/step - loss: 0.0016 - val_loss: 0.0058
Epoch 3/100
21/21 [=====] - 9s 446ms/step - loss: 0.0012 - val_loss: 0.0076
Epoch 4/100
21/21 [=====] - 9s 448ms/step - loss: 0.0011 - val_loss: 0.0045
Epoch 5/100
21/21 [=====] - 9s 445ms/step - loss: 0.0011 - val_loss: 0.0051
Epoch 6/100
21/21 [=====] - 9s 450ms/step - loss: 0.0010 - val_loss: 0.0067
Epoch 7/100
21/21 [=====] - 9s 446ms/step - loss: 9.6973e-04 - val_loss: 0.0055
Epoch 8/100
21/21 [=====] - 9s 446ms/step - loss: 9.6582e-04 - val_loss: 0.0063
Epoch 9/100
21/21 [=====] - 9s 447ms/step - loss: 9.0100e-04 - val_loss: 0.0078
Epoch 10/100
21/21 [=====] - 9s 447ms/step - loss: 8.7918e-04 - val_loss: 0.0081
Epoch 11/100
21/21 [=====] - 9s 446ms/step - loss: 8.3878e-04 - val_loss: 0.0046
Epoch 12/100

```

21/21 [=====] - 9s 446ms/step - loss: 8.0756e-04 - val\_loss: 0.0045  
Epoch 13/100  
21/21 [=====] - 9s 446ms/step - loss: 7.6312e-04 - val\_loss: 0.0070  
Epoch 14/100  
21/21 [=====] - 9s 444ms/step - loss: 7.7651e-04 - val\_loss: 0.0106  
Epoch 15/100  
21/21 [=====] - 9s 447ms/step - loss: 7.3030e-04 - val\_loss: 0.0120  
Epoch 16/100  
21/21 [=====] - 9s 445ms/step - loss: 7.2185e-04 - val\_loss: 0.0117  
Epoch 17/100  
21/21 [=====] - 9s 446ms/step - loss: 7.0774e-04 - val\_loss: 0.0109  
Epoch 18/100  
21/21 [=====] - 9s 444ms/step - loss: 6.6743e-04 - val\_loss: 0.0069  
Epoch 19/100  
21/21 [=====] - 9s 446ms/step - loss: 6.7868e-04 - val\_loss: 0.0078  
Epoch 20/100  
21/21 [=====] - 9s 445ms/step - loss: 6.8190e-04 - val\_loss: 0.0063  
Epoch 21/100  
21/21 [=====] - 9s 448ms/step - loss: 6.4858e-04 - val\_loss: 0.0091  
Epoch 22/100  
21/21 [=====] - 10s 463ms/step - loss: 6.1573e-04 - val\_loss: 0.0102  
Epoch 23/100  
21/21 [=====] - 9s 446ms/step - loss: 7.2068e-04 - val\_loss: 0.0048  
Epoch 24/100  
21/21 [=====] - 9s 430ms/step - loss: 6.1793e-04 - val\_loss: 0.0050  
Epoch 25/100  
21/21 [=====] - 9s 447ms/step - loss: 6.0539e-04 - val\_loss: 0.0055  
Epoch 26/100  
21/21 [=====] - 9s 445ms/step - loss: 5.3843e-04 - val\_loss: 0.0066  
Epoch 27/100  
21/21 [=====] - 9s 442ms/step - loss: 5.3525e-04 - val\_loss: 0.0067  
Epoch 28/100  
21/21 [=====] - 9s 444ms/step - loss: 5.1546e-04 - val\_loss: 0.0051  
Epoch 29/100  
21/21 [=====] - 9s 447ms/step - loss: 5.1158e-04 - val\_loss: 0.0056  
Epoch 30/100  
21/21 [=====] - 9s 444ms/step - loss: 4.8144e-04 - val\_loss: 0.0045  
Epoch 31/100  
21/21 [=====] - 9s 443ms/step - loss: 4.5406e-04 - val\_loss: 0.0086  
Epoch 32/100  
21/21 [=====] - 9s 437ms/step - loss: 4.9131e-04 - val\_loss: 0.0039  
Epoch 33/100

21/21 [=====] - 9s 444ms/step - loss: 4.3976e-04 - val\_loss: 0.0021  
Epoch 34/100  
21/21 [=====] - 9s 441ms/step - loss: 4.1750e-04 - val\_loss: 0.0051  
Epoch 35/100  
21/21 [=====] - 9s 443ms/step - loss: 3.7826e-04 - val\_loss: 0.0031  
Epoch 36/100  
21/21 [=====] - 9s 446ms/step - loss: 4.0094e-04 - val\_loss: 0.0043  
Epoch 37/100  
21/21 [=====] - 9s 441ms/step - loss: 3.8499e-04 - val\_loss: 0.0030  
Epoch 38/100  
21/21 [=====] - 9s 441ms/step - loss: 3.2767e-04 - val\_loss: 0.0044  
Epoch 39/100  
21/21 [=====] - 9s 447ms/step - loss: 3.1347e-04 - val\_loss: 0.0020  
Epoch 40/100  
21/21 [=====] - 9s 445ms/step - loss: 3.1695e-04 - val\_loss: 0.0017  
Epoch 41/100  
21/21 [=====] - 9s 447ms/step - loss: 3.0407e-04 - val\_loss: 0.0021  
Epoch 42/100  
21/21 [=====] - 9s 447ms/step - loss: 3.1325e-04 - val\_loss: 0.0024  
Epoch 43/100  
21/21 [=====] - 9s 439ms/step - loss: 3.8333e-04 - val\_loss: 0.0023  
Epoch 44/100  
21/21 [=====] - 9s 442ms/step - loss: 3.3562e-04 - val\_loss: 0.0037  
Epoch 45/100  
21/21 [=====] - 9s 443ms/step - loss: 2.5687e-04 - val\_loss: 0.0023  
Epoch 46/100  
21/21 [=====] - 9s 447ms/step - loss: 2.5158e-04 - val\_loss: 0.0033  
Epoch 47/100  
21/21 [=====] - 9s 453ms/step - loss: 2.4951e-04 - val\_loss: 0.0014  
Epoch 48/100  
21/21 [=====] - 9s 446ms/step - loss: 2.9410e-04 - val\_loss: 0.0042  
Epoch 49/100  
21/21 [=====] - 9s 450ms/step - loss: 2.4326e-04 - val\_loss: 0.0018  
Epoch 50/100  
21/21 [=====] - 9s 446ms/step - loss: 2.1838e-04 - val\_loss: 0.0023  
Epoch 51/100  
21/21 [=====] - 9s 446ms/step - loss: 2.0341e-04 - val\_loss: 0.0031  
Epoch 52/100  
21/21 [=====] - 9s 434ms/step - loss: 2.0791e-04 - val\_loss: 0.0011  
Epoch 53/100  
21/21 [=====] - 9s 446ms/step - loss: 1.8995e-04 - val\_loss: 0.0020  
Epoch 54/100

21/21 [=====] - 9s 444ms/step - loss: 1.8928e-04 - val\_loss: 0.0014  
Epoch 55/100  
21/21 [=====] - 9s 447ms/step - loss: 2.0658e-04 - val\_loss: 0.0013  
Epoch 56/100  
21/21 [=====] - 8s 400ms/step - loss: 1.8186e-04 - val\_loss: 0.0017  
Epoch 57/100  
21/21 [=====] - 9s 440ms/step - loss: 1.7449e-04 - val\_loss: 0.0019  
Epoch 58/100  
21/21 [=====] - 9s 446ms/step - loss: 1.5785e-04 - val\_loss: 0.0013  
Epoch 59/100  
21/21 [=====] - 9s 443ms/step - loss: 1.6955e-04 - val\_loss: 0.0010  
Epoch 60/100  
21/21 [=====] - 9s 446ms/step - loss: 1.9400e-04 - val\_loss: 0.0033  
Epoch 61/100  
21/21 [=====] - 10s 467ms/step - loss: 1.8846e-04 - val\_loss: 9.1300e-04  
Epoch 62/100  
21/21 [=====] - 9s 448ms/step - loss: 2.0368e-04 - val\_loss: 0.0012  
Epoch 63/100  
21/21 [=====] - 9s 446ms/step - loss: 1.8990e-04 - val\_loss: 0.0014  
Epoch 64/100  
21/21 [=====] - 9s 444ms/step - loss: 1.5244e-04 - val\_loss: 0.0011  
Epoch 65/100  
21/21 [=====] - 9s 448ms/step - loss: 1.6491e-04 - val\_loss: 0.0013  
Epoch 66/100  
21/21 [=====] - 9s 443ms/step - loss: 1.6471e-04 - val\_loss: 7.7323e-04  
Epoch 67/100  
21/21 [=====] - 9s 444ms/step - loss: 1.8939e-04 - val\_loss: 0.0013  
Epoch 68/100  
21/21 [=====] - 9s 450ms/step - loss: 1.4464e-04 - val\_loss: 0.0021  
Epoch 69/100  
21/21 [=====] - 9s 448ms/step - loss: 1.6168e-04 - val\_loss: 0.0011  
Epoch 70/100  
21/21 [=====] - 9s 444ms/step - loss: 1.7206e-04 - val\_loss: 8.5940e-04  
Epoch 71/100  
21/21 [=====] - 9s 440ms/step - loss: 1.6415e-04 - val\_loss: 0.0011  
Epoch 72/100  
21/21 [=====] - 9s 447ms/step - loss: 1.3930e-04 - val\_loss: 0.0022  
Epoch 73/100  
21/21 [=====] - 9s 447ms/step - loss: 2.0283e-04 - val\_loss: 0.0052  
Epoch 74/100  
21/21 [=====] - 9s 451ms/step - loss: 1.8478e-04 - val\_loss: 0.0014  
Epoch 75/100

21/21 [=====] - 9s 445ms/step - loss: 1.4276e-04 - val\_loss: 0.0024  
Epoch 76/100  
21/21 [=====] - 9s 452ms/step - loss: 1.5224e-04 - val\_loss: 0.0026  
Epoch 77/100  
21/21 [=====] - 9s 450ms/step - loss: 1.6428e-04 - val\_loss: 0.0019  
Epoch 78/100  
21/21 [=====] - 9s 448ms/step - loss: 1.3995e-04 - val\_loss: 0.0033  
Epoch 79/100  
21/21 [=====] - 9s 429ms/step - loss: 1.4810e-04 - val\_loss: 0.0021  
Epoch 80/100  
21/21 [=====] - 9s 440ms/step - loss: 1.3661e-04 - val\_loss: 0.0019  
Epoch 81/100  
21/21 [=====] - 10s 455ms/step - loss: 1.3843e-04 - val\_loss: 0.0030  
Epoch 82/100  
21/21 [=====] - 10s 456ms/step - loss: 1.6110e-04 - val\_loss: 0.0016  
Epoch 83/100  
21/21 [=====] - 9s 447ms/step - loss: 1.3818e-04 - val\_loss: 0.0023  
Epoch 84/100  
21/21 [=====] - 9s 443ms/step - loss: 1.3606e-04 - val\_loss: 0.0013  
Epoch 85/100  
21/21 [=====] - 9s 447ms/step - loss: 1.4585e-04 - val\_loss: 0.0010  
Epoch 86/100  
21/21 [=====] - 9s 448ms/step - loss: 1.3482e-04 - val\_loss: 0.0014  
Epoch 87/100  
21/21 [=====] - 9s 446ms/step - loss: 1.3320e-04 - val\_loss: 0.0013  
Epoch 88/100  
21/21 [=====] - 9s 431ms/step - loss: 1.3978e-04 - val\_loss: 0.0020  
Epoch 89/100  
21/21 [=====] - 9s 439ms/step - loss: 1.5298e-04 - val\_loss: 0.0019  
Epoch 90/100  
21/21 [=====] - 9s 447ms/step - loss: 1.8003e-04 - val\_loss: 0.0031  
Epoch 91/100  
21/21 [=====] - 9s 448ms/step - loss: 1.6370e-04 - val\_loss: 0.0015  
Epoch 92/100  
21/21 [=====] - 9s 447ms/step - loss: 1.4927e-04 - val\_loss: 0.0027  
Epoch 93/100  
21/21 [=====] - 9s 448ms/step - loss: 1.3764e-04 - val\_loss: 0.0022  
Epoch 94/100  
21/21 [=====] - 9s 446ms/step - loss: 1.3559e-04 - val\_loss: 0.0015  
Epoch 95/100  
21/21 [=====] - 9s 447ms/step - loss: 1.3769e-04 - val\_loss: 0.0029  
Epoch 96/100



```
21/21 [=====] - 9s 447ms/step - loss: 1.6741e-04 - val_loss: 0.0032
Epoch 97/100
21/21 [=====] - 9s 447ms/step - loss: 1.4076e-04 - val_loss: 0.0021
Epoch 98/100
21/21 [=====] - 9s 446ms/step - loss: 1.5296e-04 - val_loss: 0.0022
Epoch 99/100
21/21 [=====] - 9s 449ms/step - loss: 1.3798e-04 - val_loss: 0.0024
Epoch 100/100
21/21 [=====] - 9s 447ms/step - loss: 1.3930e-04 - val_loss: 0.0020
<keras.callbacks.History at 0x2d6e8dc0c40>
```

Out[42]:

```
In [43]: test_pred = sc.inverse_transform(model.predict(X_test))
```

## calc MSR

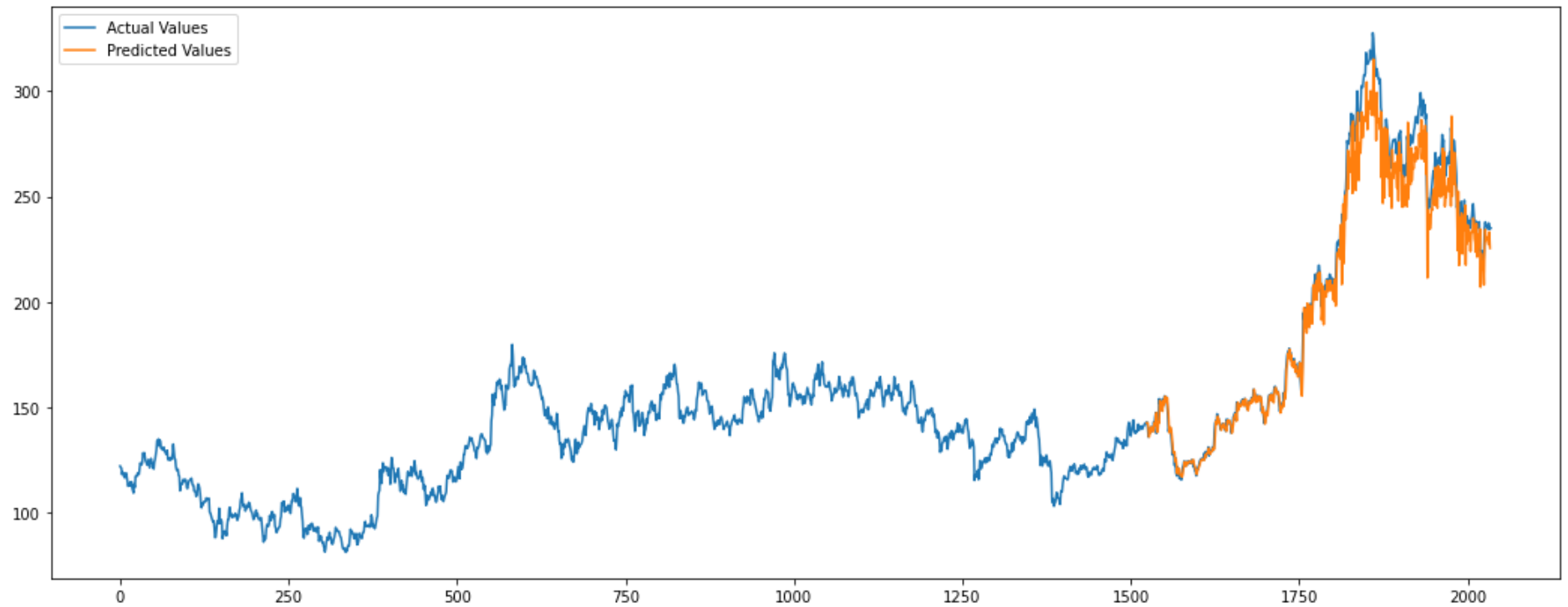
```
In [44]: from sklearn.metrics import mean_squared_error
math.sqrt(mean_squared_error(Y_test, test_pred))
```

Out[44]: 206.796146792728

## Plotting test values

```
In [48]: val = 100
testPredPlot = np.empty_like(df)
testPredPlot[:, :] = np.nan
testPredPlot[len(X_train) + (val*2) + 1:len(df) - 1, :] = test_pred
```

```
In [49]: plt.figure(figsize=(18, 7))
plt.plot(sc.inverse_transform(df), label="Actual Values")
plt.plot(testPredPlot, label="Predicted Values")
plt.legend(loc="upper left")
plt.show()
```



As you can see our test prediction matches original value as we can see... in graph above

```
In [54]: len(Y_test)
```

```
Out[54]: 510
```

```
In [77]: pred_input = Y_test[310:410].reshape(1,-1)
         pred_input.shape
```

```
Out[77]: (1, 100)
```

```
In [78]: temp = list(pred_input)
         temp = temp[0].tolist()
```

## Defining Prediction fxn

In [79]:

```
output = []
steps=100
i = 0
while (i < 30):
    if(len(temp) > 100):
        pred_input = np.array(temp[1:])
        print("Day {} input {}".format(i, pred_input))
        pred_input = pred_input.reshape(1,-1)
        pred_input = pred_input.reshape((1, steps, 1))
        pred = model.predict(pred_input, verbose = 0)
        print("Day {} output {}".format(i, pred))
        temp.extend(pred[0].tolist())
        temp = temp[1:]
        output.extend(pred.tolist())
        i = i + 1
    else:
        pred_input = pred_input.reshape((1, steps, 1))
        pred = model.predict(pred_input, verbose = 0)
        print(pred[0])
        temp.extend(pred[0].tolist())
        print(len(temp))
        output.extend(pred.tolist())
        i = i + 1

print(output)
```

[0.79536206]

101

Day 1 input [0.84763072 0.8880719 0.84497549 0.83129085 0.8306781 0.84109477  
0.85743464 0.89746732 0.89828431 0.89542484 0.90604575 0.91911765  
0.91748366 0.92177288 0.96180556 0.94750817 0.94035948 0.94056373  
0.94485294 0.94771242 0.96670752 0.95731209 0.95241013 0.94914216  
1. 0.98876634 0.96180556 0.9501634 0.91707516 0.92892157  
0.93096405 0.92422386 0.9121732 0.9121732 0.90216503 0.91115196  
0.85661765 0.84722222 0.79207516 0.78492647 0.75878268 0.81372549  
0.81495098 0.80596405 0.83394608 0.82107843 0.80923203 0.79513889  
0.76388889 0.74305556 0.75837418 0.74019608 0.77512255 0.78656046  
0.79268791 0.79473039 0.79207516 0.79534314 0.78921569 0.76776961  
0.78615196 0.76552288 0.79697712 0.8057598 0.80698529 0.81147876

0.78206699 0.73386438 0.73978758 0.74448529 0.72671569 0.72283497  
0.74040033 0.74693627 0.72630719 0.80065359 0.79187092 0.77328431  
0.78451797 0.78329248 0.80412582 0.79473039 0.7879902 0.7879902  
0.79473039 0.81168301 0.81699346 0.82986111 0.83843954 0.83884804  
0.82781863 0.82598039 0.85028595 0.85886438 0.86254085 0.88480392  
0.86519608 0.84027778 0.84497549 0.79536206]

Day 1 output [[0.6724932]]

Day 2 input [0.8880719 0.84497549 0.83129085 0.8306781 0.84109477 0.85743464  
0.89746732 0.89828431 0.89542484 0.90604575 0.91911765 0.91748366  
0.92177288 0.96180556 0.94750817 0.94035948 0.94056373 0.94485294  
0.94771242 0.96670752 0.95731209 0.95241013 0.94914216 1.  
0.98876634 0.96180556 0.9501634 0.91707516 0.92892157 0.93096405  
0.92422386 0.9121732 0.9121732 0.90216503 0.91115196 0.85661765  
0.84722222 0.79207516 0.78492647 0.75878268 0.81372549 0.81495098  
0.80596405 0.83394608 0.82107843 0.80923203 0.79513889 0.76388889  
0.74305556 0.75837418 0.74019608 0.77512255 0.78656046 0.79268791  
0.79473039 0.79207516 0.79534314 0.78921569 0.76776961 0.78615196  
0.76552288 0.79697712 0.8057598 0.80698529 0.81147876 0.78206699  
0.73386438 0.73978758 0.74448529 0.72671569 0.72283497 0.74040033  
0.74693627 0.72630719 0.80065359 0.79187092 0.77328431 0.78451797  
0.78329248 0.80412582 0.79473039 0.7879902 0.7879902 0.79473039  
0.81168301 0.81699346 0.82986111 0.83843954 0.83884804 0.82781863  
0.82598039 0.85028595 0.85886438 0.86254085 0.88480392 0.86519608  
0.84027778 0.84497549 0.79536206 0.67249322]

Day 2 output [[0.55198026]]

Day 3 input [0.84497549 0.83129085 0.8306781 0.84109477 0.85743464 0.89746732  
0.89828431 0.89542484 0.90604575 0.91911765 0.91748366 0.92177288  
0.96180556 0.94750817 0.94035948 0.94056373 0.94485294 0.94771242  
0.96670752 0.95731209 0.95241013 0.94914216 1. 0.98876634  
0.96180556 0.9501634 0.91707516 0.92892157 0.93096405 0.92422386  
0.9121732 0.9121732 0.90216503 0.91115196 0.85661765 0.84722222  
0.79207516 0.78492647 0.75878268 0.81372549 0.81495098 0.80596405  
0.83394608 0.82107843 0.80923203 0.79513889 0.76388889 0.74305556  
0.75837418 0.74019608 0.77512255 0.78656046 0.79268791 0.79473039  
0.79207516 0.79534314 0.78921569 0.76776961 0.78615196 0.76552288  
0.79697712 0.8057598 0.80698529 0.81147876 0.78206699 0.73386438  
0.73978758 0.74448529 0.72671569 0.72283497 0.74040033 0.74693627  
0.72630719 0.80065359 0.79187092 0.77328431 0.78451797 0.78329248  
0.80412582 0.79473039 0.7879902 0.7879902 0.79473039 0.81168301  
0.81699346 0.82986111 0.83843954 0.83884804 0.82781863 0.82598039  
0.85028595 0.85886438 0.86254085 0.88480392 0.86519608 0.84027778  
0.84497549 0.79536206 0.67249322 0.55198026]

Day 3 output [[0.49116346]]

Day 4 input [0.83129085 0.8306781 0.84109477 0.85743464 0.89746732 0.89828431  
0.89542484 0.90604575 0.91911765 0.91748366 0.92177288 0.96180556  
0.94750817 0.94035948 0.94056373 0.94485294 0.94771242 0.96670752  
0.95731209 0.95241013 0.94914216 1. 0.98876634 0.96180556  
0.9501634 0.91707516 0.92892157 0.93096405 0.92422386 0.9121732  
0.9121732 0.90216503 0.91115196 0.85661765 0.84722222 0.79207516  
0.78492647 0.75878268 0.81372549 0.81495098 0.80596405 0.83394608  
0.82107843 0.80923203 0.79513889 0.76388889 0.74305556 0.75837418  
0.74019608 0.77512255 0.78656046 0.79268791 0.79473039 0.79207516  
0.79534314 0.78921569 0.76776961 0.78615196 0.76552288 0.79697712  
0.8057598 0.80698529 0.81147876 0.78206699 0.73386438 0.73978758  
0.74448529 0.72671569 0.72283497 0.74040033 0.74693627 0.72630719  
0.80065359 0.79187092 0.77328431 0.78451797 0.78329248 0.80412582  
0.79473039 0.7879902 0.7879902 0.79473039 0.81168301 0.81699346  
0.82986111 0.83843954 0.83884804 0.82781863 0.82598039 0.85028595  
0.85886438 0.86254085 0.88480392 0.86519608 0.84027778 0.84497549  
0.79536206 0.67249322 0.55198026 0.49116346]

Day 4 output [[0.46929827]]

Day 5 input [0.8306781 0.84109477 0.85743464 0.89746732 0.89828431 0.89542484  
0.90604575 0.91911765 0.91748366 0.92177288 0.96180556 0.94750817  
0.94035948 0.94056373 0.94485294 0.94771242 0.96670752 0.95731209  
0.95241013 0.94914216 1. 0.98876634 0.96180556 0.9501634  
0.91707516 0.92892157 0.93096405 0.92422386 0.9121732 0.9121732  
0.90216503 0.91115196 0.85661765 0.84722222 0.79207516 0.78492647  
0.75878268 0.81372549 0.81495098 0.80596405 0.83394608 0.82107843  
0.80923203 0.79513889 0.76388889 0.74305556 0.75837418 0.74019608  
0.77512255 0.78656046 0.79268791 0.79473039 0.79207516 0.79534314  
0.78921569 0.76776961 0.78615196 0.76552288 0.79697712 0.8057598  
0.80698529 0.81147876 0.78206699 0.73386438 0.73978758 0.74448529  
0.72671569 0.72283497 0.74040033 0.74693627 0.72630719 0.80065359  
0.79187092 0.77328431 0.78451797 0.78329248 0.80412582 0.79473039  
0.7879902 0.7879902 0.79473039 0.81168301 0.81699346 0.82986111  
0.83843954 0.83884804 0.82781863 0.82598039 0.85028595 0.85886438  
0.86254085 0.88480392 0.86519608 0.84027778 0.84497549 0.79536206  
0.67249322 0.55198026 0.49116346 0.46929827]

Day 5 output [[0.45402214]]

Day 6 input [0.84109477 0.85743464 0.89746732 0.89828431 0.89542484 0.90604575  
0.91911765 0.91748366 0.92177288 0.96180556 0.94750817 0.94035948  
0.94056373 0.94485294 0.94771242 0.96670752 0.95731209 0.95241013  
0.94914216 1. 0.98876634 0.96180556 0.9501634 0.91707516  
0.92892157 0.93096405 0.92422386 0.9121732 0.9121732 0.90216503]

0.91115196 0.85661765 0.84722222 0.79207516 0.78492647 0.75878268  
0.81372549 0.81495098 0.80596405 0.83394608 0.82107843 0.80923203  
0.79513889 0.76388889 0.74305556 0.75837418 0.74019608 0.77512255  
0.78656046 0.79268791 0.79473039 0.79207516 0.79534314 0.78921569  
0.76776961 0.78615196 0.76552288 0.79697712 0.8057598 0.80698529  
0.81147876 0.78206699 0.73386438 0.73978758 0.74448529 0.72671569  
0.72283497 0.74040033 0.74693627 0.72630719 0.80065359 0.79187092  
0.77328431 0.78451797 0.78329248 0.80412582 0.79473039 0.7879902  
0.7879902 0.79473039 0.81168301 0.81699346 0.82986111 0.83843954  
0.83884804 0.82781863 0.82598039 0.85028595 0.85886438 0.86254085  
0.88480392 0.86519608 0.84027778 0.84497549 0.79536206 0.67249322  
0.55198026 0.49116346 0.46929827 0.45402214]

Day 6 output [[0.43889728]]

Day 7 input [0.85743464 0.89746732 0.89828431 0.89542484 0.90604575 0.91911765

0.91748366 0.92177288 0.96180556 0.94750817 0.94035948 0.94056373  
0.94485294 0.94771242 0.96670752 0.95731209 0.95241013 0.94914216  
1. 0.98876634 0.96180556 0.9501634 0.91707516 0.92892157  
0.93096405 0.92422386 0.9121732 0.9121732 0.90216503 0.91115196  
0.85661765 0.84722222 0.79207516 0.78492647 0.75878268 0.81372549  
0.81495098 0.80596405 0.83394608 0.82107843 0.80923203 0.79513889  
0.76388889 0.74305556 0.75837418 0.74019608 0.77512255 0.78656046  
0.79268791 0.79473039 0.79207516 0.79534314 0.78921569 0.76776961  
0.78615196 0.76552288 0.79697712 0.8057598 0.80698529 0.81147876  
0.78206699 0.73386438 0.73978758 0.74448529 0.72671569 0.72283497  
0.74040033 0.74693627 0.72630719 0.80065359 0.79187092 0.77328431  
0.78451797 0.78329248 0.80412582 0.79473039 0.7879902 0.7879902  
0.79473039 0.81168301 0.81699346 0.82986111 0.83843954 0.83884804  
0.82781863 0.82598039 0.85028595 0.85886438 0.86254085 0.88480392  
0.86519608 0.84027778 0.84497549 0.79536206 0.67249322 0.55198026  
0.49116346 0.46929827 0.45402214 0.43889728]

Day 7 output [[0.4289838]]

Day 8 input [0.89746732 0.89828431 0.89542484 0.90604575 0.91911765 0.91748366

0.92177288 0.96180556 0.94750817 0.94035948 0.94056373 0.94485294  
0.94771242 0.96670752 0.95731209 0.95241013 0.94914216 1.  
0.98876634 0.96180556 0.9501634 0.91707516 0.92892157 0.93096405  
0.92422386 0.9121732 0.9121732 0.90216503 0.91115196 0.85661765  
0.84722222 0.79207516 0.78492647 0.75878268 0.81372549 0.81495098  
0.80596405 0.83394608 0.82107843 0.80923203 0.79513889 0.76388889  
0.74305556 0.75837418 0.74019608 0.77512255 0.78656046 0.79268791  
0.79473039 0.79207516 0.79534314 0.78921569 0.76776961 0.78615196  
0.76552288 0.79697712 0.8057598 0.80698529 0.81147876 0.78206699  
0.73386438 0.73978758 0.74448529 0.72671569 0.72283497 0.74040033

0.74693627 0.72630719 0.80065359 0.79187092 0.77328431 0.78451797  
0.78329248 0.80412582 0.79473039 0.7879902 0.7879902 0.79473039  
0.81168301 0.81699346 0.82986111 0.83843954 0.83884804 0.82781863  
0.82598039 0.85028595 0.85886438 0.86254085 0.88480392 0.86519608  
0.84027778 0.84497549 0.79536206 0.67249322 0.55198026 0.49116346  
0.46929827 0.45402214 0.43889728 0.42898381]

Day 8 output [[0.42534664]]

Day 9 input [0.89828431 0.89542484 0.90604575 0.91911765 0.91748366 0.92177288  
0.96180556 0.94750817 0.94035948 0.94056373 0.94485294 0.94771242  
0.96670752 0.95731209 0.95241013 0.94914216 1. 0.98876634  
0.96180556 0.9501634 0.91707516 0.92892157 0.93096405 0.92422386  
0.9121732 0.9121732 0.90216503 0.91115196 0.85661765 0.84722222  
0.79207516 0.78492647 0.75878268 0.81372549 0.81495098 0.80596405  
0.83394608 0.82107843 0.80923203 0.79513889 0.76388889 0.74305556  
0.75837418 0.74019608 0.77512255 0.78656046 0.79268791 0.79473039  
0.79207516 0.79534314 0.78921569 0.76776961 0.78615196 0.76552288  
0.79697712 0.8057598 0.80698529 0.81147876 0.78206699 0.73386438  
0.73978758 0.74448529 0.72671569 0.72283497 0.74040033 0.74693627  
0.72630719 0.80065359 0.79187092 0.77328431 0.78451797 0.78329248  
0.80412582 0.79473039 0.7879902 0.7879902 0.79473039 0.81168301  
0.81699346 0.82986111 0.83843954 0.83884804 0.82781863 0.82598039  
0.85028595 0.85886438 0.86254085 0.88480392 0.86519608 0.84027778  
0.84497549 0.79536206 0.67249322 0.55198026 0.49116346 0.46929827  
0.45402214 0.43889728 0.42898381 0.42534664]

Day 9 output [[0.42578852]]

Day 10 input [0.89542484 0.90604575 0.91911765 0.91748366 0.92177288 0.96180556  
0.94750817 0.94035948 0.94056373 0.94485294 0.94771242 0.96670752  
0.95731209 0.95241013 0.94914216 1. 0.98876634 0.96180556  
0.9501634 0.91707516 0.92892157 0.93096405 0.92422386 0.9121732  
0.9121732 0.90216503 0.91115196 0.85661765 0.84722222 0.79207516  
0.78492647 0.75878268 0.81372549 0.81495098 0.80596405 0.83394608  
0.82107843 0.80923203 0.79513889 0.76388889 0.74305556 0.75837418  
0.74019608 0.77512255 0.78656046 0.79268791 0.79473039 0.79207516  
0.79534314 0.78921569 0.76776961 0.78615196 0.76552288 0.79697712  
0.8057598 0.80698529 0.81147876 0.78206699 0.73386438 0.73978758  
0.74448529 0.72671569 0.72283497 0.74040033 0.74693627 0.72630719  
0.80065359 0.79187092 0.77328431 0.78451797 0.78329248 0.80412582  
0.79473039 0.7879902 0.7879902 0.79473039 0.81168301 0.81699346  
0.82986111 0.83843954 0.83884804 0.82781863 0.82598039 0.85028595  
0.85886438 0.86254085 0.88480392 0.86519608 0.84027778 0.84497549  
0.79536206 0.67249322 0.55198026 0.49116346 0.46929827 0.45402214  
0.43889728 0.42898381 0.42534664 0.42578852]

Day 10 output [[0.4284406]]

Day 11 input [0.90604575 0.91911765 0.91748366 0.92177288 0.96180556 0.94750817  
0.94035948 0.94056373 0.94485294 0.94771242 0.96670752 0.95731209  
0.95241013 0.94914216 1. 0.98876634 0.96180556 0.9501634  
0.91707516 0.92892157 0.93096405 0.92422386 0.9121732 0.9121732  
0.90216503 0.91115196 0.85661765 0.84722222 0.79207516 0.78492647  
0.75878268 0.81372549 0.81495098 0.80596405 0.83394608 0.82107843  
0.80923203 0.79513889 0.76388889 0.74305556 0.75837418 0.74019608  
0.77512255 0.78656046 0.79268791 0.79473039 0.79207516 0.79534314  
0.78921569 0.76776961 0.78615196 0.76552288 0.79697712 0.8057598  
0.80698529 0.81147876 0.78206699 0.73386438 0.73978758 0.74448529  
0.72671569 0.72283497 0.74040033 0.74693627 0.72630719 0.80065359  
0.79187092 0.77328431 0.78451797 0.78329248 0.80412582 0.79473039  
0.7879902 0.7879902 0.79473039 0.81168301 0.81699346 0.82986111  
0.83843954 0.83884804 0.82781863 0.82598039 0.85028595 0.85886438  
0.86254085 0.88480392 0.86519608 0.84027778 0.84497549 0.79536206  
0.67249322 0.55198026 0.49116346 0.46929827 0.45402214 0.43889728  
0.42898381 0.42534664 0.42578852 0.4284406 ]

Day 11 output [[0.43218586]]

Day 12 input [0.91911765 0.91748366 0.92177288 0.96180556 0.94750817 0.94035948  
0.94056373 0.94485294 0.94771242 0.96670752 0.95731209 0.95241013  
0.94914216 1. 0.98876634 0.96180556 0.9501634 0.91707516  
0.92892157 0.93096405 0.92422386 0.9121732 0.9121732 0.90216503  
0.91115196 0.85661765 0.84722222 0.79207516 0.78492647 0.75878268  
0.81372549 0.81495098 0.80596405 0.83394608 0.82107843 0.80923203  
0.79513889 0.76388889 0.74305556 0.75837418 0.74019608 0.77512255  
0.78656046 0.79268791 0.79473039 0.79207516 0.79534314 0.78921569  
0.76776961 0.78615196 0.76552288 0.79697712 0.8057598 0.80698529  
0.81147876 0.78206699 0.73386438 0.73978758 0.74448529 0.72671569  
0.72283497 0.74040033 0.74693627 0.72630719 0.80065359 0.79187092  
0.77328431 0.78451797 0.78329248 0.80412582 0.79473039 0.7879902  
0.7879902 0.79473039 0.81168301 0.81699346 0.82986111 0.83843954  
0.83884804 0.82781863 0.82598039 0.85028595 0.85886438 0.86254085  
0.88480392 0.86519608 0.84027778 0.84497549 0.79536206 0.67249322  
0.55198026 0.49116346 0.46929827 0.45402214 0.43889728 0.42898381  
0.42534664 0.42578852 0.4284406 0.43218586]

Day 12 output [[0.43614265]]

Day 13 input [0.91748366 0.92177288 0.96180556 0.94750817 0.94035948 0.94056373  
0.94485294 0.94771242 0.96670752 0.95731209 0.95241013 0.94914216  
1. 0.98876634 0.96180556 0.9501634 0.91707516 0.92892157  
0.93096405 0.92422386 0.9121732 0.9121732 0.90216503 0.91115196  
0.85661765 0.84722222 0.79207516 0.78492647 0.75878268 0.81372549



0.81495098 0.80596405 0.83394608 0.82107843 0.80923203 0.79513889  
0.76388889 0.74305556 0.75837418 0.74019608 0.77512255 0.78656046  
0.79268791 0.79473039 0.79207516 0.79534314 0.78921569 0.76776961  
0.78615196 0.76552288 0.79697712 0.8057598 0.80698529 0.81147876  
0.78206699 0.73386438 0.73978758 0.74448529 0.72671569 0.72283497  
0.74040033 0.74693627 0.72630719 0.80065359 0.79187092 0.77328431  
0.78451797 0.78329248 0.80412582 0.79473039 0.7879902 0.7879902  
0.79473039 0.81168301 0.81699346 0.82986111 0.83843954 0.83884804  
0.82781863 0.82598039 0.85028595 0.85886438 0.86254085 0.88480392  
0.86519608 0.84027778 0.84497549 0.79536206 0.67249322 0.55198026  
0.49116346 0.46929827 0.45402214 0.43889728 0.42898381 0.42534664  
0.42578852 0.4284406 0.43218586 0.43614265]

Day 13 output [[0.4395612]]

Day 14 input [0.92177288 0.96180556 0.94750817 0.94035948 0.94056373 0.94485294  
0.94771242 0.96670752 0.95731209 0.95241013 0.94914216 1.  
0.98876634 0.96180556 0.9501634 0.91707516 0.92892157 0.93096405  
0.92422386 0.9121732 0.9121732 0.90216503 0.91115196 0.85661765  
0.84722222 0.79207516 0.78492647 0.75878268 0.81372549 0.81495098  
0.80596405 0.83394608 0.82107843 0.80923203 0.79513889 0.76388889  
0.74305556 0.75837418 0.74019608 0.77512255 0.78656046 0.79268791  
0.79473039 0.79207516 0.79534314 0.78921569 0.76776961 0.78615196  
0.76552288 0.79697712 0.8057598 0.80698529 0.81147876 0.78206699  
0.73386438 0.73978758 0.74448529 0.72671569 0.72283497 0.74040033  
0.74693627 0.72630719 0.80065359 0.79187092 0.77328431 0.78451797  
0.78329248 0.80412582 0.79473039 0.7879902 0.7879902 0.79473039  
0.81168301 0.81699346 0.82986111 0.83843954 0.83884804 0.82781863  
0.82598039 0.85028595 0.85886438 0.86254085 0.88480392 0.86519608  
0.84027778 0.84497549 0.79536206 0.67249322 0.55198026 0.49116346  
0.46929827 0.45402214 0.43889728 0.42898381 0.42534664 0.42578852  
0.4284406 0.43218586 0.43614265 0.43956119]

Day 14 output [[0.44189227]]

Day 15 input [0.96180556 0.94750817 0.94035948 0.94056373 0.94485294 0.94771242  
0.96670752 0.95731209 0.95241013 0.94914216 1. 0.98876634  
0.96180556 0.9501634 0.91707516 0.92892157 0.93096405 0.92422386  
0.9121732 0.9121732 0.90216503 0.91115196 0.85661765 0.84722222  
0.79207516 0.78492647 0.75878268 0.81372549 0.81495098 0.80596405  
0.83394608 0.82107843 0.80923203 0.79513889 0.76388889 0.74305556  
0.75837418 0.74019608 0.77512255 0.78656046 0.79268791 0.79473039  
0.79207516 0.79534314 0.78921569 0.76776961 0.78615196 0.76552288  
0.79697712 0.8057598 0.80698529 0.81147876 0.78206699 0.73386438  
0.73978758 0.74448529 0.72671569 0.72283497 0.74040033 0.74693627  
0.72630719 0.80065359 0.79187092 0.77328431 0.78451797 0.78329248

0.80412582 0.79473039 0.7879902 0.7879902 0.79473039 0.81168301  
0.81699346 0.82986111 0.83843954 0.83884804 0.82781863 0.82598039  
0.85028595 0.85886438 0.86254085 0.88480392 0.86519608 0.84027778  
0.84497549 0.79536206 0.67249322 0.55198026 0.49116346 0.46929827  
0.45402214 0.43889728 0.42898381 0.42534664 0.42578852 0.4284406  
0.43218586 0.43614265 0.43956119 0.44189227]

Day 15 output [[0.44279933]]

Day 16 input [0.94750817 0.94035948 0.94056373 0.94485294 0.94771242 0.96670752  
0.95731209 0.95241013 0.94914216 1. 0.98876634 0.96180556  
0.9501634 0.91707516 0.92892157 0.93096405 0.92422386 0.9121732  
0.9121732 0.90216503 0.91115196 0.85661765 0.84722222 0.79207516  
0.78492647 0.75878268 0.81372549 0.81495098 0.80596405 0.83394608  
0.82107843 0.80923203 0.79513889 0.76388889 0.74305556 0.75837418  
0.74019608 0.77512255 0.78656046 0.79268791 0.79473039 0.79207516  
0.79534314 0.78921569 0.76776961 0.78615196 0.76552288 0.79697712  
0.8057598 0.80698529 0.81147876 0.78206699 0.73386438 0.73978758  
0.74448529 0.72671569 0.72283497 0.74040033 0.74693627 0.72630719  
0.80065359 0.79187092 0.77328431 0.78451797 0.78329248 0.80412582  
0.79473039 0.7879902 0.7879902 0.79473039 0.81168301 0.81699346  
0.82986111 0.83843954 0.83884804 0.82781863 0.82598039 0.85028595  
0.85886438 0.86254085 0.88480392 0.86519608 0.84027778 0.84497549  
0.79536206 0.67249322 0.55198026 0.49116346 0.46929827 0.45402214  
0.43889728 0.42898381 0.42534664 0.42578852 0.4284406 0.43218586  
0.43614265 0.43956119 0.44189227 0.44279933]

Day 16 output [[0.44213253]]

Day 17 input [0.94035948 0.94056373 0.94485294 0.94771242 0.96670752 0.95731209  
0.95241013 0.94914216 1. 0.98876634 0.96180556 0.9501634  
0.91707516 0.92892157 0.93096405 0.92422386 0.9121732 0.9121732  
0.90216503 0.91115196 0.85661765 0.84722222 0.79207516 0.78492647  
0.75878268 0.81372549 0.81495098 0.80596405 0.83394608 0.82107843  
0.80923203 0.79513889 0.76388889 0.74305556 0.75837418 0.74019608  
0.77512255 0.78656046 0.79268791 0.79473039 0.79207516 0.79534314  
0.78921569 0.76776961 0.78615196 0.76552288 0.79697712 0.8057598  
0.80698529 0.81147876 0.78206699 0.73386438 0.73978758 0.74448529  
0.72671569 0.72283497 0.74040033 0.74693627 0.72630719 0.80065359  
0.79187092 0.77328431 0.78451797 0.78329248 0.80412582 0.79473039  
0.7879902 0.7879902 0.79473039 0.81168301 0.81699346 0.82986111  
0.83843954 0.83884804 0.82781863 0.82598039 0.85028595 0.85886438  
0.86254085 0.88480392 0.86519608 0.84027778 0.84497549 0.79536206  
0.67249322 0.55198026 0.49116346 0.46929827 0.45402214 0.43889728  
0.42898381 0.42534664 0.42578852 0.4284406 0.43218586 0.43614265  
0.43956119 0.44189227 0.44279933 0.44213253]

Day 17 output [[0.43990052]]

Day 18 input [0.94056373 0.94485294 0.94771242 0.96670752 0.95731209 0.95241013

0.94914216 1. 0.98876634 0.96180556 0.9501634 0.91707516  
0.92892157 0.93096405 0.92422386 0.9121732 0.9121732 0.90216503  
0.91115196 0.85661765 0.84722222 0.79207516 0.78492647 0.75878268  
0.81372549 0.81495098 0.80596405 0.83394608 0.82107843 0.80923203  
0.79513889 0.76388889 0.74305556 0.75837418 0.74019608 0.77512255  
0.78656046 0.79268791 0.79473039 0.79207516 0.79534314 0.78921569  
0.76776961 0.78615196 0.76552288 0.79697712 0.8057598 0.80698529  
0.81147876 0.78206699 0.73386438 0.73978758 0.74448529 0.72671569  
0.72283497 0.74040033 0.74693627 0.72630719 0.80065359 0.79187092  
0.77328431 0.78451797 0.78329248 0.80412582 0.79473039 0.7879902  
0.7879902 0.79473039 0.81168301 0.81699346 0.82986111 0.83843954  
0.83884804 0.82781863 0.82598039 0.85028595 0.85886438 0.86254085  
0.88480392 0.86519608 0.84027778 0.84497549 0.79536206 0.67249322  
0.55198026 0.49116346 0.46929827 0.45402214 0.43889728 0.42898381  
0.42534664 0.42578852 0.4284406 0.43218586 0.43614265 0.43956119  
0.44189227 0.44279933 0.44213253 0.43990052]

Day 18 output [[0.4362419]]

Day 19 input [0.94485294 0.94771242 0.96670752 0.95731209 0.95241013 0.94914216

1. 0.98876634 0.96180556 0.9501634 0.91707516 0.92892157  
0.93096405 0.92422386 0.9121732 0.9121732 0.90216503 0.91115196  
0.85661765 0.84722222 0.79207516 0.78492647 0.75878268 0.81372549  
0.81495098 0.80596405 0.83394608 0.82107843 0.80923203 0.79513889  
0.76388889 0.74305556 0.75837418 0.74019608 0.77512255 0.78656046  
0.79268791 0.79473039 0.79207516 0.79534314 0.78921569 0.76776961  
0.78615196 0.76552288 0.79697712 0.8057598 0.80698529 0.81147876  
0.78206699 0.73386438 0.73978758 0.74448529 0.72671569 0.72283497  
0.74040033 0.74693627 0.72630719 0.80065359 0.79187092 0.77328431  
0.78451797 0.78329248 0.80412582 0.79473039 0.7879902 0.7879902  
0.79473039 0.81168301 0.81699346 0.82986111 0.83843954 0.83884804  
0.82781863 0.82598039 0.85028595 0.85886438 0.86254085 0.88480392  
0.86519608 0.84027778 0.84497549 0.79536206 0.67249322 0.55198026  
0.49116346 0.46929827 0.45402214 0.43889728 0.42898381 0.42534664  
0.42578852 0.4284406 0.43218586 0.43614265 0.43956119 0.44189227  
0.44279933 0.44213253 0.43990052 0.43624189]

Day 19 output [[0.4313935]]

Day 20 input [0.94771242 0.96670752 0.95731209 0.95241013 0.94914216 1.

0.98876634 0.96180556 0.9501634 0.91707516 0.92892157 0.93096405  
0.92422386 0.9121732 0.9121732 0.90216503 0.91115196 0.85661765  
0.84722222 0.79207516 0.78492647 0.75878268 0.81372549 0.81495098  
0.80596405 0.83394608 0.82107843 0.80923203 0.79513889 0.76388889

0.74305556 0.75837418 0.74019608 0.77512255 0.78656046 0.79268791  
0.79473039 0.79207516 0.79534314 0.78921569 0.76776961 0.78615196  
0.76552288 0.79697712 0.8057598 0.80698529 0.81147876 0.78206699  
0.73386438 0.73978758 0.74448529 0.72671569 0.72283497 0.74040033  
0.74693627 0.72630719 0.80065359 0.79187092 0.77328431 0.78451797  
0.78329248 0.80412582 0.79473039 0.7879902 0.7879902 0.79473039  
0.81168301 0.81699346 0.82986111 0.83843954 0.83884804 0.82781863  
0.82598039 0.85028595 0.85886438 0.86254085 0.88480392 0.86519608  
0.84027778 0.84497549 0.79536206 0.67249322 0.55198026 0.49116346  
0.46929827 0.45402214 0.43889728 0.42898381 0.42534664 0.42578852  
0.4284406 0.43218586 0.43614265 0.43956119 0.44189227 0.44279933  
0.44213253 0.43990052 0.43624189 0.4313935 ]

Day 20 output [[0.42565528]]

Day 21 input [0.96670752 0.95731209 0.95241013 0.94914216 1. 0.98876634

0.96180556 0.9501634 0.91707516 0.92892157 0.93096405 0.92422386  
0.9121732 0.9121732 0.90216503 0.91115196 0.85661765 0.84722222  
0.79207516 0.78492647 0.75878268 0.81372549 0.81495098 0.80596405  
0.83394608 0.82107843 0.80923203 0.79513889 0.76388889 0.74305556  
0.75837418 0.74019608 0.77512255 0.78656046 0.79268791 0.79473039  
0.79207516 0.79534314 0.78921569 0.76776961 0.78615196 0.76552288  
0.79697712 0.8057598 0.80698529 0.81147876 0.78206699 0.73386438  
0.73978758 0.74448529 0.72671569 0.72283497 0.74040033 0.74693627  
0.72630719 0.80065359 0.79187092 0.77328431 0.78451797 0.78329248  
0.80412582 0.79473039 0.7879902 0.7879902 0.79473039 0.81168301  
0.81699346 0.82986111 0.83843954 0.83884804 0.82781863 0.82598039  
0.85028595 0.85886438 0.86254085 0.88480392 0.86519608 0.84027778  
0.84497549 0.79536206 0.67249322 0.55198026 0.49116346 0.46929827  
0.45402214 0.43889728 0.42898381 0.42534664 0.42578852 0.4284406  
0.43218586 0.43614265 0.43956119 0.44189227 0.44279933 0.44213253  
0.43990052 0.43624189 0.4313935 0.42565528]

Day 21 output [[0.41935417]]

Day 22 input [0.95731209 0.95241013 0.94914216 1. 0.98876634 0.96180556

0.9501634 0.91707516 0.92892157 0.93096405 0.92422386 0.9121732  
0.9121732 0.90216503 0.91115196 0.85661765 0.84722222 0.79207516  
0.78492647 0.75878268 0.81372549 0.81495098 0.80596405 0.83394608  
0.82107843 0.80923203 0.79513889 0.76388889 0.74305556 0.75837418  
0.74019608 0.77512255 0.78656046 0.79268791 0.79473039 0.79207516  
0.79534314 0.78921569 0.76776961 0.78615196 0.76552288 0.79697712  
0.8057598 0.80698529 0.81147876 0.78206699 0.73386438 0.73978758  
0.74448529 0.72671569 0.72283497 0.74040033 0.74693627 0.72630719  
0.80065359 0.79187092 0.77328431 0.78451797 0.78329248 0.80412582  
0.79473039 0.7879902 0.7879902 0.79473039 0.81168301 0.81699346

0.82986111 0.83843954 0.83884804 0.82781863 0.82598039 0.85028595  
0.85886438 0.86254085 0.88480392 0.86519608 0.84027778 0.84497549  
0.79536206 0.67249322 0.55198026 0.49116346 0.46929827 0.45402214  
0.43889728 0.42898381 0.42534664 0.42578852 0.4284406 0.43218586  
0.43614265 0.43956119 0.44189227 0.44279933 0.44213253 0.43990052  
0.43624189 0.4313935 0.42565528 0.41935417]

Day 22 output [[0.41280964]]

Day 23 input [0.95241013 0.94914216 1. 0.98876634 0.96180556 0.9501634  
0.91707516 0.92892157 0.93096405 0.92422386 0.9121732 0.9121732  
0.90216503 0.91115196 0.85661765 0.84722222 0.79207516 0.78492647  
0.75878268 0.81372549 0.81495098 0.80596405 0.83394608 0.82107843  
0.80923203 0.79513889 0.76388889 0.74305556 0.75837418 0.74019608  
0.77512255 0.78656046 0.79268791 0.79473039 0.79207516 0.79534314  
0.78921569 0.76776961 0.78615196 0.76552288 0.79697712 0.8057598  
0.80698529 0.81147876 0.78206699 0.73386438 0.73978758 0.74448529  
0.72671569 0.72283497 0.74040033 0.74693627 0.72630719 0.80065359  
0.79187092 0.77328431 0.78451797 0.78329248 0.80412582 0.79473039  
0.7879902 0.7879902 0.79473039 0.81168301 0.81699346 0.82986111  
0.83843954 0.83884804 0.82781863 0.82598039 0.85028595 0.85886438  
0.86254085 0.88480392 0.86519608 0.84027778 0.84497549 0.79536206  
0.67249322 0.55198026 0.49116346 0.46929827 0.45402214 0.43889728  
0.42898381 0.42534664 0.42578852 0.4284406 0.43218586 0.43614265  
0.43956119 0.44189227 0.44279933 0.44213253 0.43990052 0.43624189  
0.4313935 0.42565528 0.41935417 0.41280964]

Day 23 output [[0.40630594]]

Day 24 input [0.94914216 1. 0.98876634 0.96180556 0.9501634 0.91707516  
0.92892157 0.93096405 0.92422386 0.9121732 0.9121732 0.90216503  
0.91115196 0.85661765 0.84722222 0.79207516 0.78492647 0.75878268  
0.81372549 0.81495098 0.80596405 0.83394608 0.82107843 0.80923203  
0.79513889 0.76388889 0.74305556 0.75837418 0.74019608 0.77512255  
0.78656046 0.79268791 0.79473039 0.79207516 0.79534314 0.78921569  
0.76776961 0.78615196 0.76552288 0.79697712 0.8057598 0.80698529  
0.81147876 0.78206699 0.73386438 0.73978758 0.74448529 0.72671569  
0.72283497 0.74040033 0.74693627 0.72630719 0.80065359 0.79187092  
0.77328431 0.78451797 0.78329248 0.80412582 0.79473039 0.7879902  
0.7879902 0.79473039 0.81168301 0.81699346 0.82986111 0.83843954  
0.83884804 0.82781863 0.82598039 0.85028595 0.85886438 0.86254085  
0.88480392 0.86519608 0.84027778 0.84497549 0.79536206 0.67249322  
0.55198026 0.49116346 0.46929827 0.45402214 0.43889728 0.42898381  
0.42534664 0.42578852 0.4284406 0.43218586 0.43614265 0.43956119  
0.44189227 0.44279933 0.44213253 0.43990052 0.43624189 0.4313935  
0.42565528 0.41935417 0.41280964 0.40630594]

Day 24 output [[0.4000724]]

Day 25 input [1. 0.98876634 0.96180556 0.9501634 0.91707516 0.92892157  
0.93096405 0.92422386 0.9121732 0.9121732 0.90216503 0.91115196  
0.85661765 0.84722222 0.79207516 0.78492647 0.75878268 0.81372549  
0.81495098 0.80596405 0.83394608 0.82107843 0.80923203 0.79513889  
0.76388889 0.74305556 0.75837418 0.74019608 0.77512255 0.78656046  
0.79268791 0.79473039 0.79207516 0.79534314 0.78921569 0.76776961  
0.78615196 0.76552288 0.79697712 0.8057598 0.80698529 0.81147876  
0.78206699 0.73386438 0.73978758 0.74448529 0.72671569 0.72283497  
0.74040033 0.74693627 0.72630719 0.80065359 0.79187092 0.77328431  
0.78451797 0.78329248 0.80412582 0.79473039 0.7879902 0.7879902  
0.79473039 0.81168301 0.81699346 0.82986111 0.83843954 0.83884804  
0.82781863 0.82598039 0.85028595 0.85886438 0.86254085 0.88480392  
0.86519608 0.84027778 0.84497549 0.79536206 0.67249322 0.55198026  
0.49116346 0.46929827 0.45402214 0.43889728 0.42898381 0.42534664  
0.42578852 0.4284406 0.43218586 0.43614265 0.43956119 0.44189227  
0.44279933 0.44213253 0.43990052 0.43624189 0.4313935 0.42565528  
0.41935417 0.41280964 0.40630594 0.4000724 ]

Day 25 output [[0.39427316]]

Day 26 input [0.98876634 0.96180556 0.9501634 0.91707516 0.92892157 0.93096405  
0.92422386 0.9121732 0.9121732 0.90216503 0.91115196 0.85661765  
0.84722222 0.79207516 0.78492647 0.75878268 0.81372549 0.81495098  
0.80596405 0.83394608 0.82107843 0.80923203 0.79513889 0.76388889  
0.74305556 0.75837418 0.74019608 0.77512255 0.78656046 0.79268791  
0.79473039 0.79207516 0.79534314 0.78921569 0.76776961 0.78615196  
0.76552288 0.79697712 0.8057598 0.80698529 0.81147876 0.78206699  
0.73386438 0.73978758 0.74448529 0.72671569 0.72283497 0.74040033  
0.74693627 0.72630719 0.80065359 0.79187092 0.77328431 0.78451797  
0.78329248 0.80412582 0.79473039 0.7879902 0.7879902 0.79473039  
0.81168301 0.81699346 0.82986111 0.83843954 0.83884804 0.82781863  
0.82598039 0.85028595 0.85886438 0.86254085 0.88480392 0.86519608  
0.84027778 0.84497549 0.79536206 0.67249322 0.55198026 0.49116346  
0.46929827 0.45402214 0.43889728 0.42898381 0.42534664 0.42578852  
0.4284406 0.43218586 0.43614265 0.43956119 0.44189227 0.44279933  
0.44213253 0.43990052 0.43624189 0.4313935 0.42565528 0.41935417  
0.41280964 0.40630594 0.4000724 0.39427316]

Day 26 output [[0.38900536]]

Day 27 input [0.96180556 0.9501634 0.91707516 0.92892157 0.93096405 0.92422386  
0.9121732 0.9121732 0.90216503 0.91115196 0.85661765 0.84722222  
0.79207516 0.78492647 0.75878268 0.81372549 0.81495098 0.80596405  
0.83394608 0.82107843 0.80923203 0.79513889 0.76388889 0.74305556  
0.75837418 0.74019608 0.77512255 0.78656046 0.79268791 0.79473039

0.79207516 0.79534314 0.78921569 0.76776961 0.78615196 0.76552288  
0.79697712 0.8057598 0.80698529 0.81147876 0.78206699 0.73386438  
0.73978758 0.74448529 0.72671569 0.72283497 0.74040033 0.74693627  
0.72630719 0.80065359 0.79187092 0.77328431 0.78451797 0.78329248  
0.80412582 0.79473039 0.7879902 0.7879902 0.79473039 0.81168301  
0.81699346 0.82986111 0.83843954 0.83884804 0.82781863 0.82598039  
0.85028595 0.85886438 0.86254085 0.88480392 0.86519608 0.84027778  
0.84497549 0.79536206 0.67249322 0.55198026 0.49116346 0.46929827  
0.45402214 0.43889728 0.42898381 0.42534664 0.42578852 0.4284406  
0.43218586 0.43614265 0.43956119 0.44189227 0.44279933 0.44213253  
0.43990052 0.43624189 0.4313935 0.42565528 0.41935417 0.41280964  
0.40630594 0.4000724 0.39427316 0.38900536]

Day 27 output [[0.3843052]]

Day 28 input [0.9501634 0.91707516 0.92892157 0.93096405 0.92422386 0.9121732

0.9121732 0.90216503 0.91115196 0.85661765 0.84722222 0.79207516  
0.78492647 0.75878268 0.81372549 0.81495098 0.80596405 0.83394608  
0.82107843 0.80923203 0.79513889 0.76388889 0.74305556 0.75837418  
0.74019608 0.77512255 0.78656046 0.79268791 0.79473039 0.79207516  
0.79534314 0.78921569 0.76776961 0.78615196 0.76552288 0.79697712  
0.8057598 0.80698529 0.81147876 0.78206699 0.73386438 0.73978758  
0.74448529 0.72671569 0.72283497 0.74040033 0.74693627 0.72630719  
0.80065359 0.79187092 0.77328431 0.78451797 0.78329248 0.80412582  
0.79473039 0.7879902 0.7879902 0.79473039 0.81168301 0.81699346  
0.82986111 0.83843954 0.83884804 0.82781863 0.82598039 0.85028595  
0.85886438 0.86254085 0.88480392 0.86519608 0.84027778 0.84497549  
0.79536206 0.67249322 0.55198026 0.49116346 0.46929827 0.45402214  
0.43889728 0.42898381 0.42534664 0.42578852 0.4284406 0.43218586  
0.43614265 0.43956119 0.44189227 0.44279933 0.44213253 0.43990052  
0.43624189 0.4313935 0.42565528 0.41935417 0.41280964 0.40630594  
0.4000724 0.39427316 0.38900536 0.38430521]

Day 28 output [[0.38015825]]

Day 29 input [0.91707516 0.92892157 0.93096405 0.92422386 0.9121732 0.9121732

0.90216503 0.91115196 0.85661765 0.84722222 0.79207516 0.78492647  
0.75878268 0.81372549 0.81495098 0.80596405 0.83394608 0.82107843  
0.80923203 0.79513889 0.76388889 0.74305556 0.75837418 0.74019608  
0.77512255 0.78656046 0.79268791 0.79473039 0.79207516 0.79534314  
0.78921569 0.76776961 0.78615196 0.76552288 0.79697712 0.8057598  
0.80698529 0.81147876 0.78206699 0.73386438 0.73978758 0.74448529  
0.72671569 0.72283497 0.74040033 0.74693627 0.72630719 0.80065359  
0.79187092 0.77328431 0.78451797 0.78329248 0.80412582 0.79473039  
0.7879902 0.7879902 0.79473039 0.81168301 0.81699346 0.82986111  
0.83843954 0.83884804 0.82781863 0.82598039 0.85028595 0.85886438

```
0.86254085 0.88480392 0.86519608 0.84027778 0.84497549 0.79536206
0.67249322 0.55198026 0.49116346 0.46929827 0.45402214 0.43889728
0.42898381 0.42534664 0.42578852 0.4284406 0.43218586 0.43614265
0.43956119 0.44189227 0.44279933 0.44213253 0.43990052 0.43624189
0.4313935 0.42565528 0.41935417 0.41280964 0.40630594 0.4000724
0.39427316 0.38900536 0.38430521 0.38015825]
```

```
Day 29 output [[0.37651244]]
```

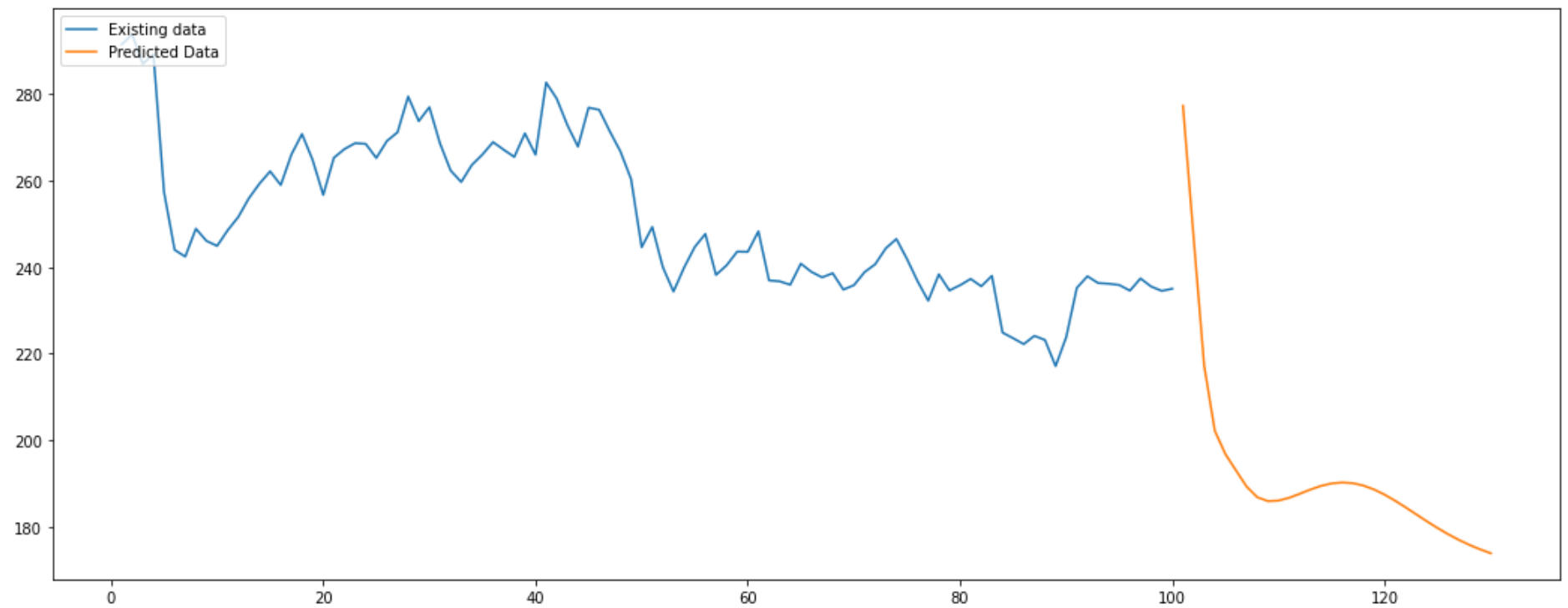
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[0.3801582455635071], [0.37651243805885315]]
```

## predicting future stock

```
In [80]: new_day = np.arange(1, 101)
pred_day = np.arange(101, 131)
```

```
In [81]: plt.figure(figsize=(18, 7))
plt.plot(new_day, sc.inverse_transform(df[1935:]), label="Existing data")
plt.plot(pred_day, sc.inverse_transform(output), label="Predicted Data")
plt.legend(loc="upper left")
plt.show()
```





Thank you

In [ ]: